

Lecture-L1. 生物信息学导论

本章内容

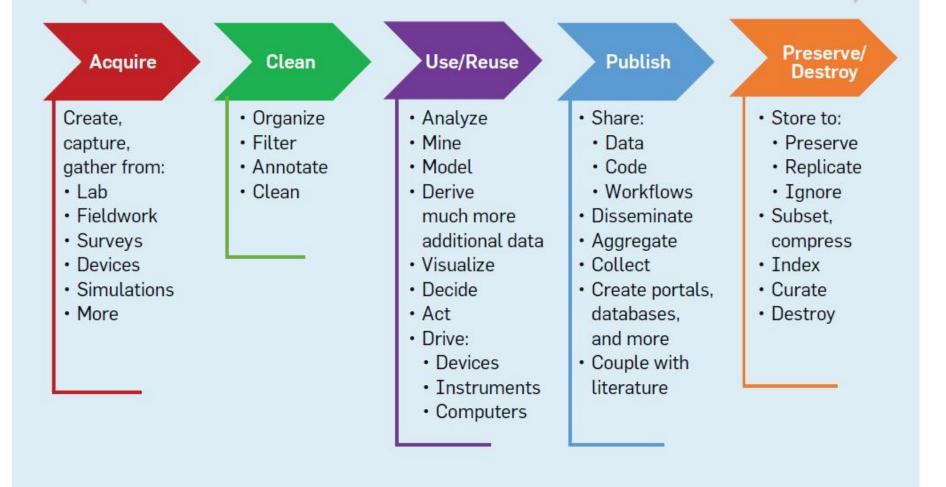
□1.1 引言--数据的价值 □1.2 什么是生物信息学? □1.3 生物信息学发展简史 □1.4 生物信息学的应用领域 □1.5 大数据时代的生物信息学 □1.6 总结与展望

🖡 ^{第1节}:引言—数据的价值

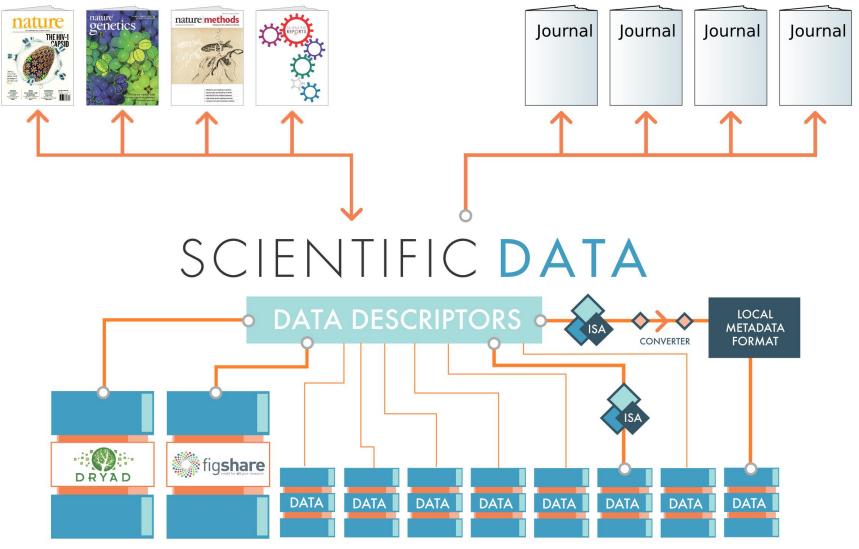


Varied Big Data in Our Life

{Ethics, Policy, Regulatory, Stewardship, Platform, Domain} Environment

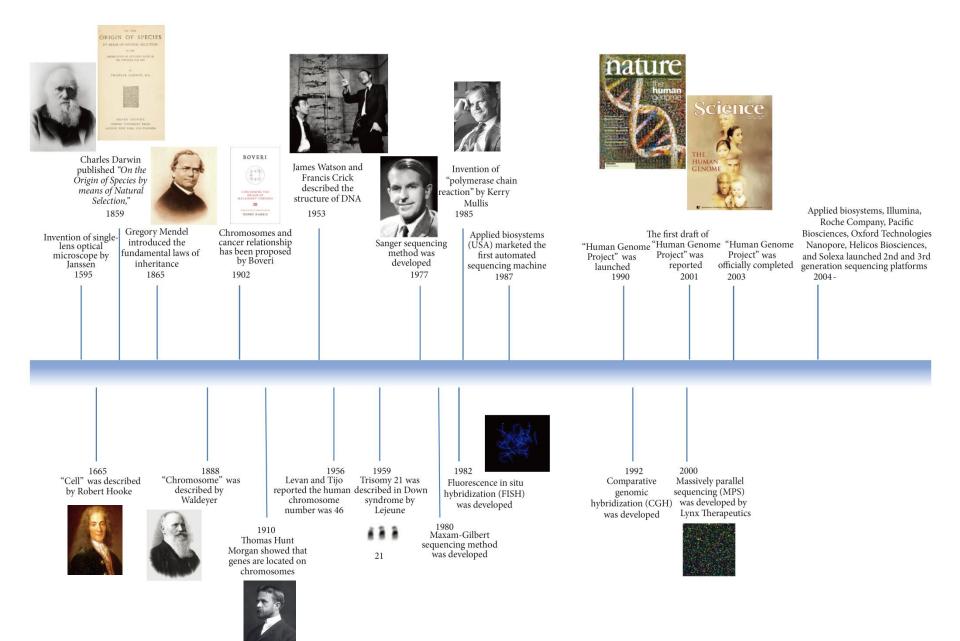


Framework used in processing Big Data



Ref: http://blogs.nature.com/

Scientific Data to complement and promote public data repositories



| | flower color | flowerposition | seed color | seed shape | pod shape | pod color | stem length |
|------------------------|------------------|----------------|-------------|------------------|--------------------|------------|--|
| | purple | axial | yellow | round | inflated | green | 👝 tall |
| Р | × | × | × | × | × ÷ | × | XXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXX |
| | white | terminal | green | wrinkled | constricted | yellow | dwarf |
| F1 | | - | \bigcirc | \bigcirc | | | A CONTRACTOR |
| | purple | axial | yellow | round | inflated | green | P tall |
| F ₁ Parents | со ригрlе | axial | o yellow | round | inflated | green | tall |
| | 705 purple | 651 axial | 6022 yellow | 5474 round | 882 inflated | 428 green | 787 tall |
| Fz Phenotype | | | | | | | A CONTRACTOR |
| | 224 white | 207 terminal | 2001 green | 1850 wrinkled | 299 constricted | 152 yellow | 277 dwarf |
| ratio | 3.15 : 1 | 3.14 : 1 | 2.82:1 | 2.96:1 | 2.95:1 | 3.01:1 | 2.84:1 |

Versuche Thanzon Hybriden Grager Mentel. [Hergeley in Sun Biflingun min 8. former 38. mary 1865] Enlistence Bemerkungen

tingthap bapangtungan ustep an Jangtonya sinthalt and yananana dia Pananta manana dia biakan termakan ya nyada nanana dia Pananta fatta ya sin tan tengahan Bagdanityyad anya dia tenga tenta yaka di palanan Bagdanityyad and nation daptan yaka di palanan di pala di pala and nation daptan yaka di palanan di pala di pala and nation daptan yaka pala di pananakan pala di pala sin tananjan ja natana tengananakan mana tengaha si na ja nafagan

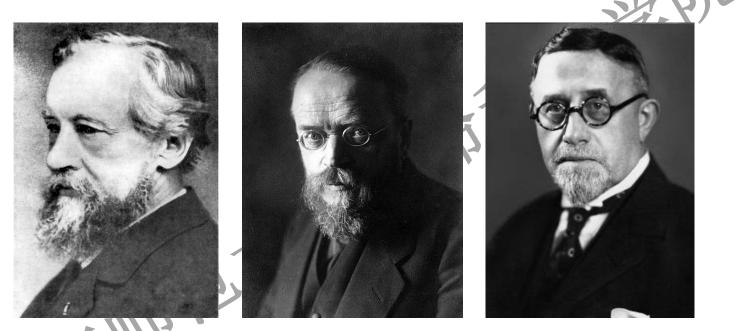
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The first page of the monuscript of Mendel's Experiments in Plant Hybridization, which was published in 1865.





Rediscovery of Mendel's work



Hugo de Vries (Netherlands)

Carl Correns (*Germany*) Erich Tschermak (*Austria*)



英勇的人民子弟兵在抢救地震中的受难群众(2008.05.12)

第 21卷第 3期 2006年 9月

 33°

基于可公度方法的川滇地区地震趋势研究

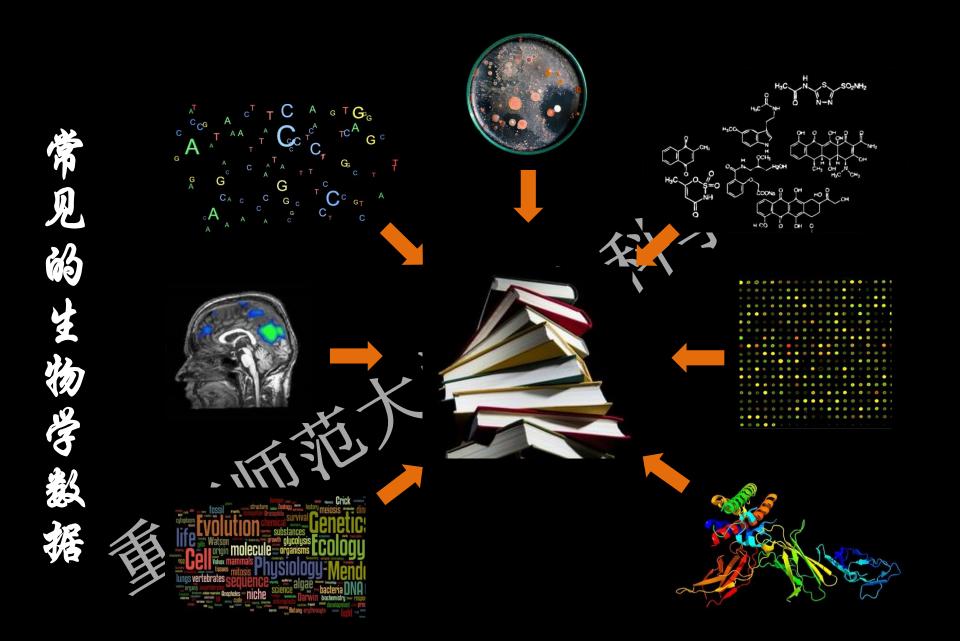
龙小霞, 延军平, 孙虎, 王祖正

(陕西师范大学 旅游与环境学院,陕西 西安 710062)

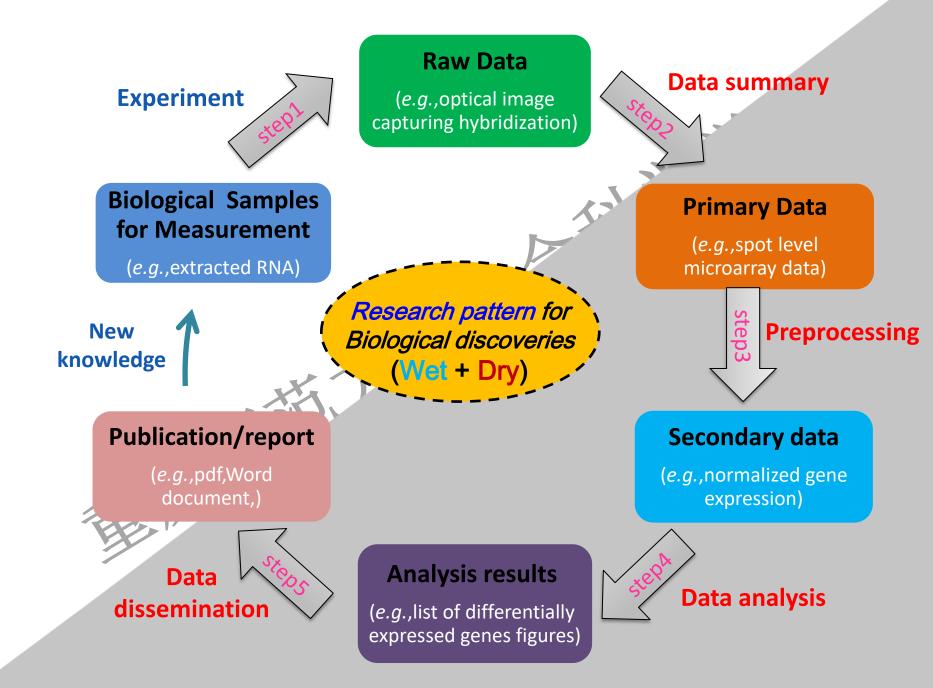
摘要:川滇地区为我国大陆最显著的强震活动区域,地震活动频繁。在对川滇地区强震灾害数据分析的基础上,应

数据是极其有用的,值得挖掘!





Bioinformatics: from data to knowledge



Biology, experimental or computational?



WORKING LIFE

By Elisabeth Pain



Biology, wet and dry

arah Teichmann's work on how cells regulate gene expression and build protein complexes recently won her a European Molecular Biology Organization Gold Medal. At 40, Teichmann holds a joint appointment with the European Bioinformatics Institute and the Wellcome Trust Sanger Institute in Hinxton, U.K. She leads a systems biology group of 17 researchers that uses both computational methods and lab experimentation. *Science* Careers asked Teichmann how she combines the two approaches. This interview has been edited for clarity and brevity.



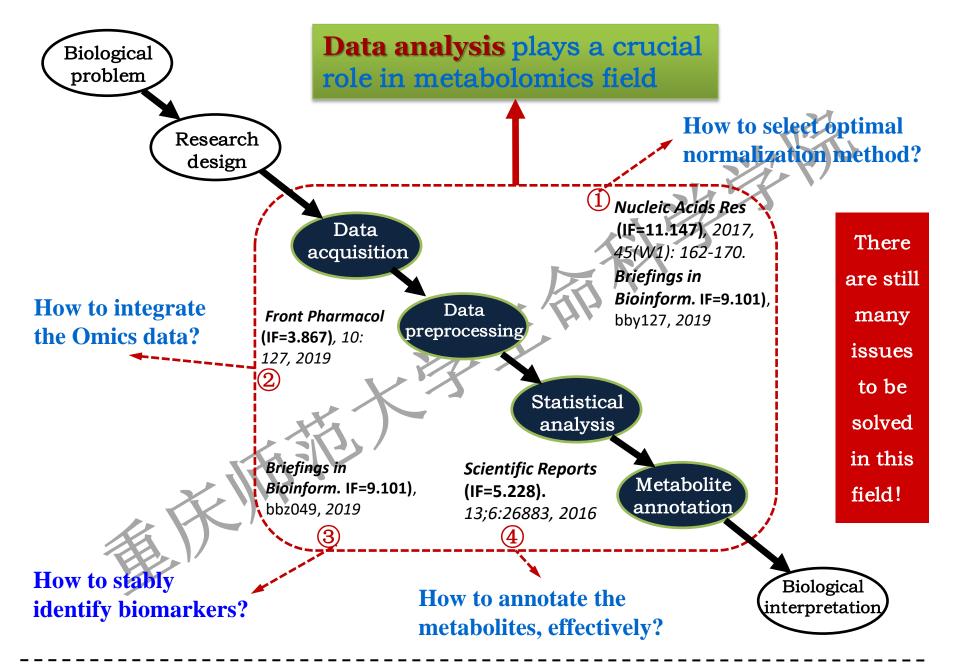
"What unifies both wet and dry work is the conceptual part of the science."

Q: Was computational biology a risky career choice?

A: Yes. But I never looked back, even though at one point I came to feel that computational biology and bioinformatics were viewed as eccentric and unorthodox.

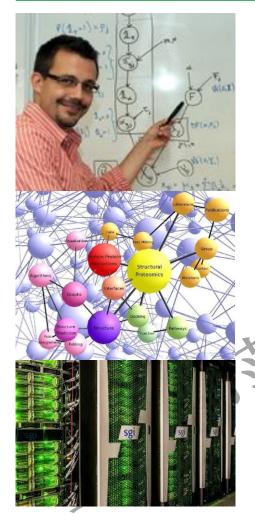
My *Ph.D.* mentor exuded such unwavering optimism and confidence (坚定不移的乐观和信心), however, that it made his lab a great place to work. <u>Altogether, during my *Ph.D.*, I published 10 papers.</u>

Ref: Teichmann, S. & Pain, E. Biology, wet and dry. Science 349, 662 (2015).



Li, B. *et al.* Performance evaluation and online realization of data-driven normalization methods used in LC/MS based untargeted metabolomics analysis. *Scientific reports* **6**, 38881 (2016).





All biology is computational biology

- Here, I argue that computational thinking and techniques are so central to the quest of understanding life that today all biology is computational biology.
- Computational biology brings order into our understanding of life, it makes biological concepts rigorous and testable, and it provides a reference map that holds together individual insights.
- The next modern synthesis in biology will be driven by mathematical, statistical, and computational methods being absorbed into mainstream biological training, turning biology into a quantitative science.

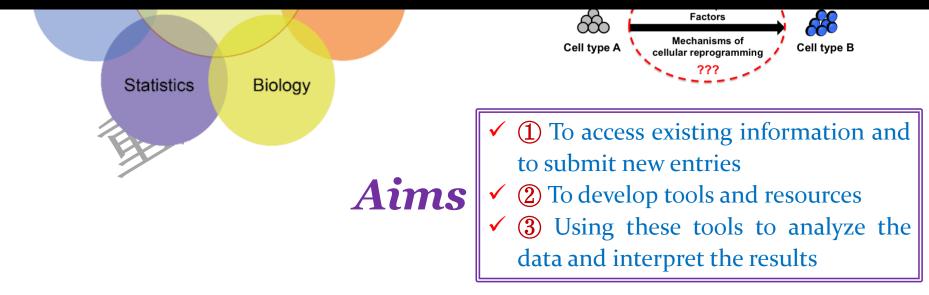
Ref: Pain, E. Biology, wet and dry. *PLoS Biology* **349**, 662-662 (2017).

★ 第2节:什么是生物信息学

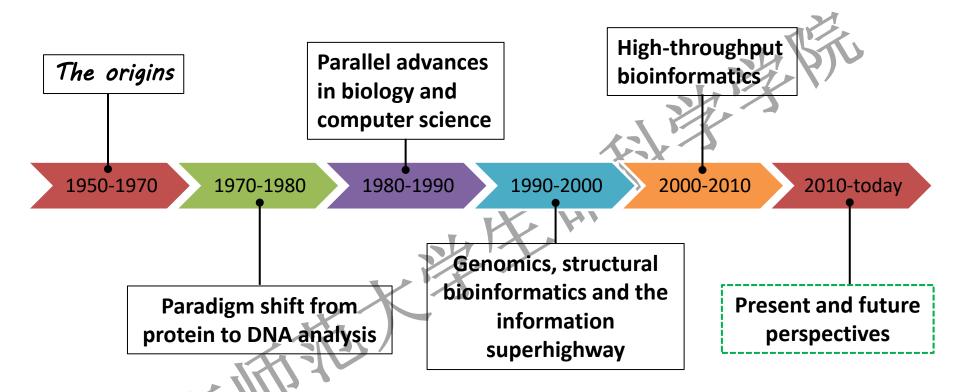
Bioinformatics is an interdisciplinary field, and it mainly develops methods and software tools for understanding biological data.



生物信息学是研究生物医学资源中蕴含的重要信息的学科,其<u>核心是解决生物学</u> 问题,常规的研究内容包括生物大分子的序列、结构和功能,以及它们之间的相 互作用等。



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It is easy for researchers to believe that modern bioinformatics are relatively recent, coming to the rescue of NGS data analysis. However, the very beginnings of bioinformatics occurred more than 50 years ago, when desktop computers were still a hypothesis and DNA could not yet be sequenced.

THE RELATIVE SIZE **OF PARTICLES**

From the COVID-19 pandemic to the U.S. West Coast wildfires, some of the biggest threats now are also the most microscopic.

A particle needs to be 10 microns (μ m) or less before it can be inhaled into your respiratory tract. But just how small are these specks?

Here's a look at the relative sizes of some familiar particles ¥

HUMAN HAIR 50-180µm > FOR SCALE

FINE BEACH SAND 90µm >

GRAIN OF SALT 60 mm

WHITE BLOOD CELL 25µm >

GRAIN OF POLLEN 15µm >

DUST PARTICLE (PM10) <10 µm >

RED BLOOD CELL 7-8µm 7

RESPIRATORY DROPLETS 5-10 µm >

DUST PARTICLE (PM2.5) 2.5µm >

BACTERIUM 1-3µm > WILDFIRE SMOKE 0.4-0.7 µm > CORONAVIRUS 0.1-0.5µm × T4 BACTERIOPHAGE 0.225µm ZIKA VIRUS 0.045µm >

Sing

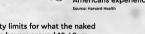
Wildfire smoke can persist in the air for

Pollen can trigger allergic reactions and hay fever—which 1 in 5 Americans experience every year.

The visibility limits for what the naked eye can see hovers around 10-40µm.

to carry smaller particles within them, such as dust or coronavirus.

several days, and even months.



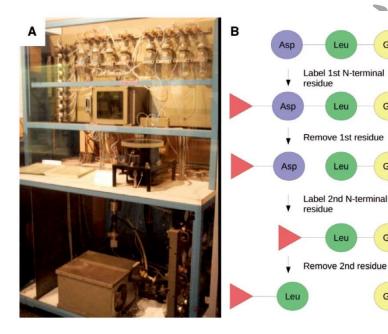
Respiratory droplets have the potential

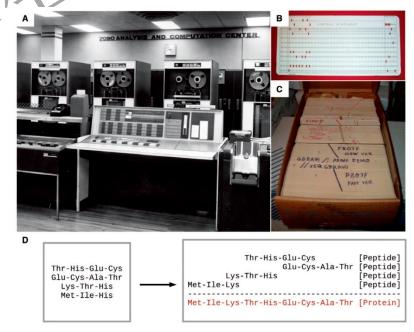
| | 1 euro col | in H | luman hai | В | acteria |] [| particles | Small mo | nm-materials | Multi-scales |
|--------------------|------------|----------------------------------|---------------------------|-------------|------------------------------|--------------------|------------------|------------------|---------------------|----------------------|
| Largest 1 1m | 0.1 1 | 0 ⁻² 10 ⁻³ | 10 ⁻⁴ 100μm | | 10 ⁻⁶ μm | 10 ⁻⁷ | 10 ⁻⁸ | 10 ⁻⁹ | 10 m 110 m 1Å | from micro |
| A volle | \supset | naterials | | Red blood c | | Nano Transistor | | DNA | Atoms | to macro |

| Symbol | Name | Factor | Symbol | Name | Factor |
|--------|-------|------------------|--------|-------|-------------------|
| Y | yotta | 10 ²⁴ | У | yokto | 10 ⁻²⁴ |
| Z | zetta | 10 ²¹ | z | zepto | 10 ⁻²¹ |
| Е | exa | 10 ¹⁸ | а | atto | 10 ⁻¹⁸ |
| Р | peta | 10 ¹⁵ | f | femto | 10 ⁻¹⁵ |
| Т | tera | 10 ¹² | р | pico | 10 ⁻¹² |
| G | giga | 10 ⁹ | n | nano | 10 ⁻⁹ |
| М | mega | 10 ⁶ | μ | micro | 10 ⁻⁶ |
| k | kilo | 10 ³ | m | milli | 10 ⁻³ |
| h | hecto | 10 ² | С | centi | 10 ⁻² |
| da | deka | 10 ¹ | d | deci | 10 ⁻¹ |

1950~1970: The origins

- It did not start with DNA analysis
- Protein analysis was the starting point
- Dayhoff: the first bioinformatician
- The computer-assisted genealogy of life (生命谱系)
- A mathematical framework for amino acid substitutions





Edman Sequencing

COMPROTEIN

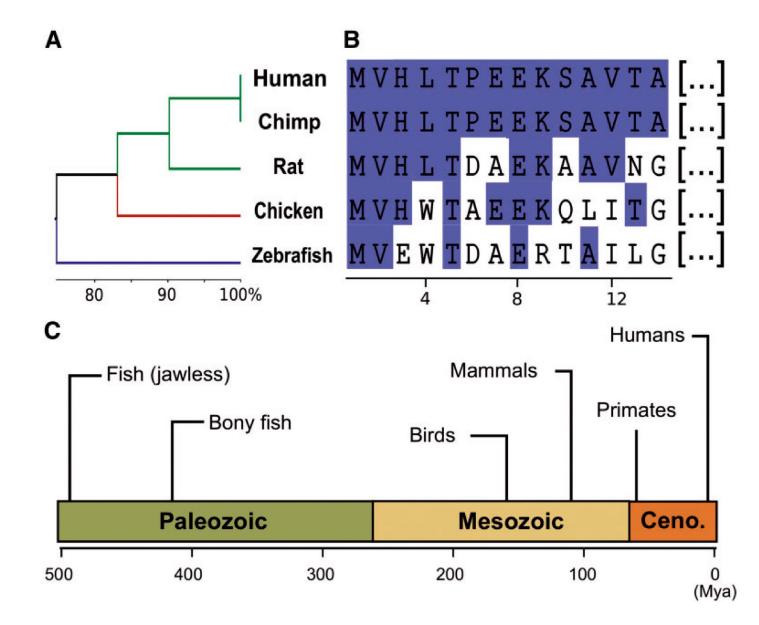
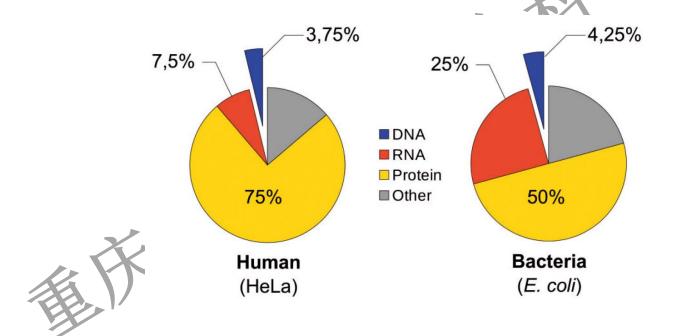


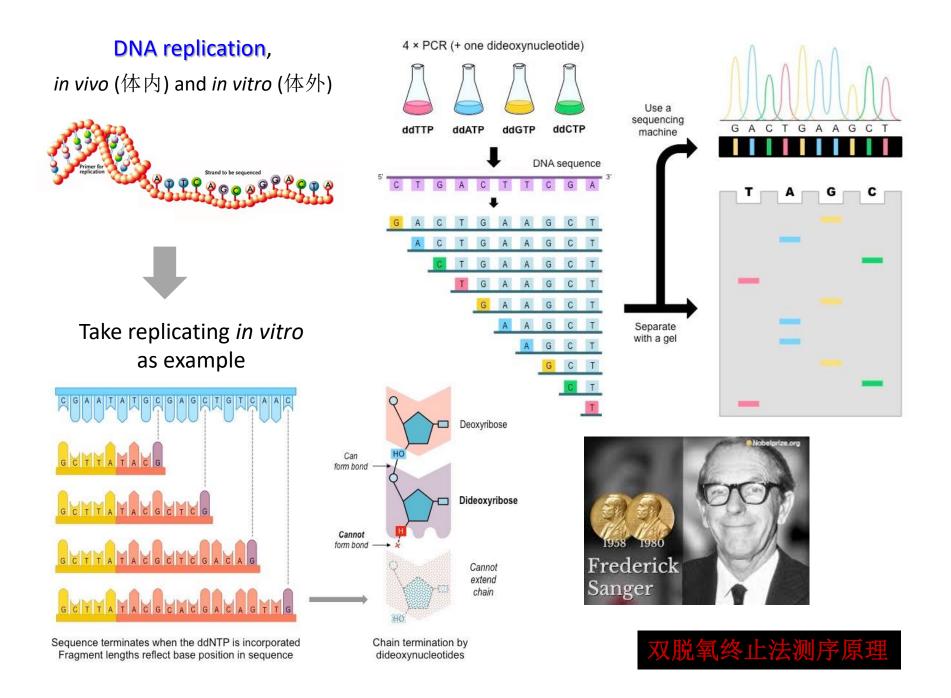
Figure 3. Sequence dissimilarity between orthologous proteins from model organisms correlates with their evolutionary history as evidenced by the fossil record. (A) Average distance tree of hemoglobin subunit beta-1 (HBB-1) from human (Homo sapiens), chimpanzee (Pan troglodytes), rat (Rattus norvegicus), chicken (Gallus gallus) and zebrafish (Danio rerio). (B) Alignment view of the first 14 amino acid residues of HBB-1 compared in (A) (residues highlighted in blue are identical to the human HBB-1 sequence). (C) Timeline of earliest fossils found for different aquatic and terrestrial animals.

1970~1980: Paradigm shift from protein to DNA analysis

- Deciphering of the DNA language: the genetic code
- Cost-efficient reading of DNA
- Using DNA sequences in phylogenetic inference



DNA is the least abundant macromolecular cell component that can be sequenced.



1980~1990: Parallel advances in biology and computer science

- Molecular methods to target and amplify specific genes;
- Access to computers and specialized software
- Bioinformatics and the free software movement
- Desktop computers and new programming languages





DEC ODO-8, 1965 a 'minicomputer' fairly had the dimensions and weight of a small household refrigerator

DEC VAX-11/780 Minicomputer. From right to left: The computer module, two tape storage units, a monitor and a terminal. The GCG software package was initially designed to run on this computer.



| Software | Year released | Use | Reference |
|------------------------|------------------|--|--------------|
| GeneQuiz | 1994 (oldest) | Workbench for protein sequence analysis | [65] |
| LabBase | 1998 | Making relational databases of sequence data | [66] |
| Phred-Phrap- Consed | 1998 | Genome assembly and finishing | [67] |
| Swissknife MUMmer | 1999 1999 | Parsing of SWISS-PROT data Whole genome alignment | [68] [69] |

HP-9000 desktop workstation running the Unix-based system HP-UX. Image: Thomas Schanz//CC-BY-SA 3.0.

PubMed Key: (perl bioinformatics) AND ("1987"[Date-Publication]: "2000" [Date-Publication]).

Table 3. Notable nonscripting and/or statistical programming languages used in bioinformatics

| | Fortran ^a | С | | R | Java |
|--|--|------------------------------------|-----------|--|---|
| First appeared | 1957 | 1972 | - | 1993 | 1995 |
| Typical use | Algorithmics, calculations, programming modules for other applications | Optimized command-line tools | = | Statistical analysis, data visualization | Graphical user interfaces, data visualization, network analysis |
| Notable fields of application | Biochemistry, Structural Bioinformatics | Various | nguage (: | Metagenomics, Transcriptomics, Systems Biology | Genomics, Proteomics, Systems Biology |
| Specialized bioinformatics repository? | None | None | (1987) | Bioconductor, [73], since 2002 | BioJava [74], since 2002 |
| Example software or packages | Clustal [32, 33], WHAT IF [75] | MUSCLE [76], PhyloBayes [77] |] | edgeR [78], phyloseq [79] | Jalview [80], Jemboss [81], Cytoscape [82] |

1990~2000: Genomics, bioinformatics and the information superhighway

| | | In 1995, the first co a free-living organi | ism (<i>Haemophi</i> | lus influenzae) | |
|--|--|---|---|--|--|
| Dawn of the genor Bioinformatics onli | was sequenced. However, the turning point that started the genomic era, as we know it actually, was the publication of the human genome at the beginning of the 21st century. | | | | |
| Beyond sequence a | analysis: st | ructural bioinf | formatics | | |
| A Hierarchical shotgun sequencing (NIH approach) | | e shotgun sequencing era approach) | Hierarch | ical shotgun | |
| Genomic DNA Construct large insert library (~175 kb) | Genomic DNA Construct shotgun library (inserts 3-5 kb) | | whole ge shotgun | <i>sequencing</i> . proaches | |
| Determine minimal set covering whole genome Shotgun Assembly CTGGCGGTGGCGCTGGTGAACATTGTGGAACGCAGCATTACCTAT | Assembly using paired-end data | Sequence insert from both ends | exemplif methodo betweer the publ private (| fied the ological rivalry ic (<i>NIH, A</i>) and <i>Celera, B</i>) o sequence the | |

Genome sequence

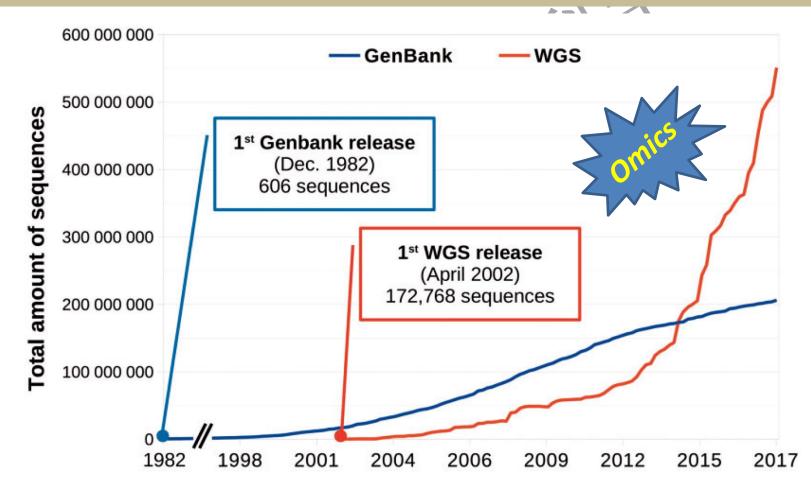
Genome sequence

2000~2010: High-throughput bioinformatics

Second-generation sequencing

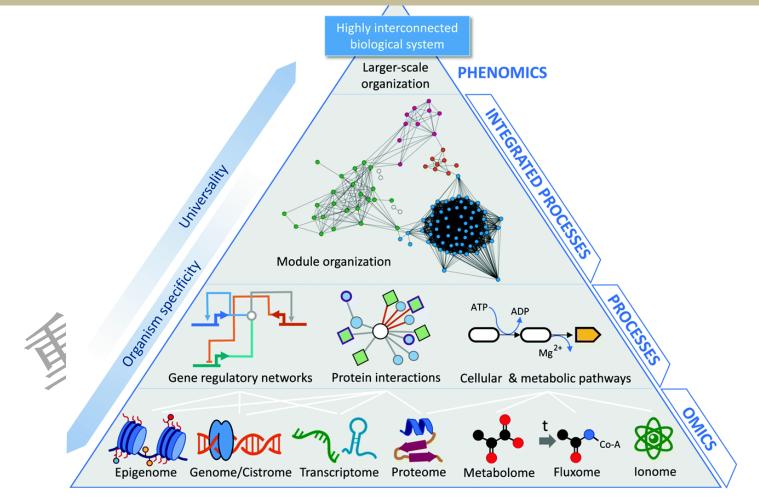
Biological Big Data

High-performance bioinformatics and collaborative computing



2010~Today: Present and future perspectives

- Clearly defining the bioinformatician profession
- ▶ Is the term 'bioinformatics' now obsolete (过时的)?
- Towards modeling life as a whole: systems biology



Indeed, the use of computers has become ubiquitous in biology, as well as in most natural sciences (physics,

chemistry, mathematics, cryptography, etc.),

but interestingly,

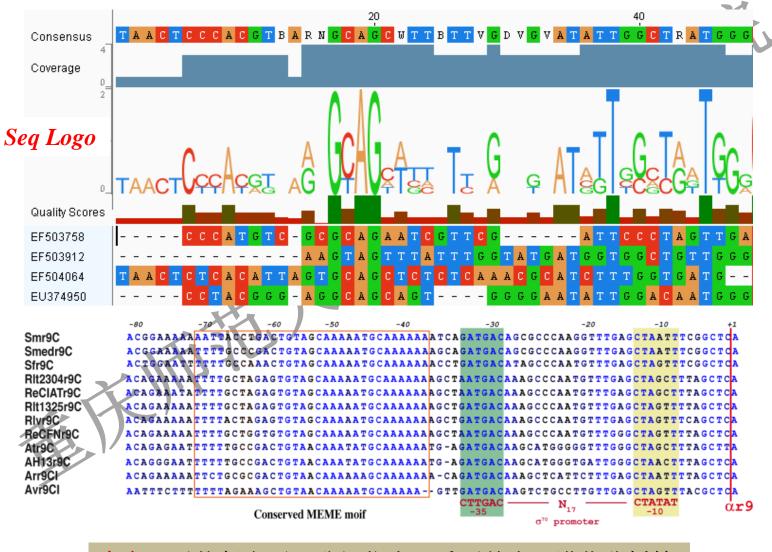
only biology has a specific term to refer to the use of computers in this discipline ('bioinformatics').

Why is that so?

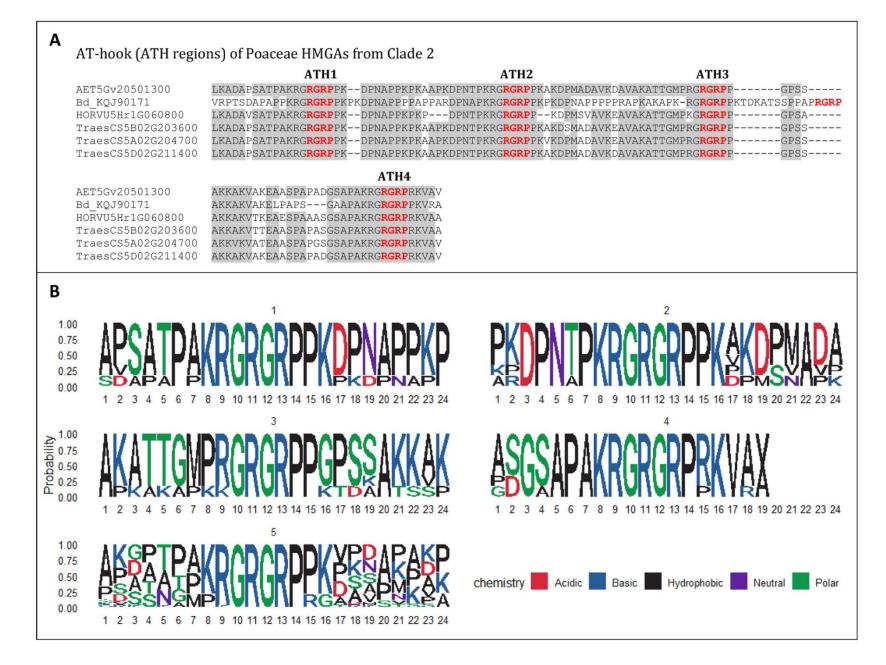


重要知识点 ✓ 序列比对、装配 ✓ 基于预测 结构预测 ✓ 分子间相互作用

◆ 4.1 序列比对



意义:寻找保守区,酶切位点,重要基序,进化分析等



Ref: Negi A P, et al. Computational biology and chemistry, 2020, 87: 107306.

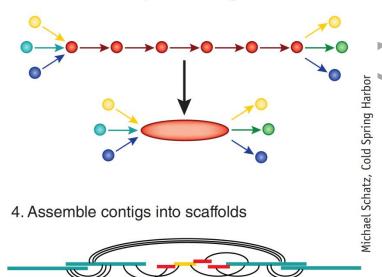
4.2 序列装配(组装)

Ref: Baker M. *Nature methods*, 2012, 9(4): 333-337.

1. Fragment DNA and sequence



- 2. Find overlaps between reads
- ...AGCCTAGACCTACAGGATGCGCGACACGT GGATGCGCGACACGTCGCATATCCGGT....
- 3. Assemble overlaps into contigs



Genome assembly stitches together a genome from short sequenced pieces of DNA.

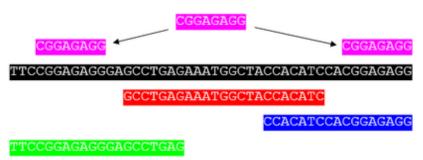
De novo assembly

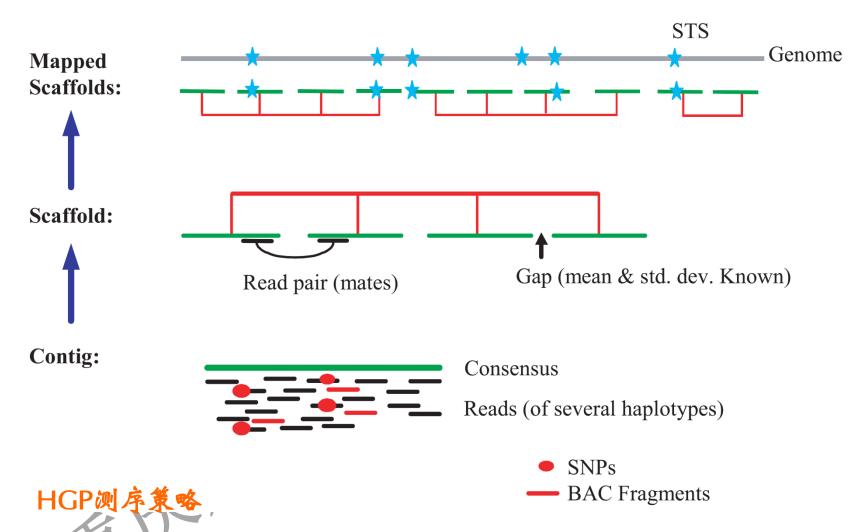
- Genome Assembly Create new reference 'from scratch'
- Examine reads for overlapping sequence
- Contig longer assembled sequence from short reads
 - Scaffold assembled contigs

Chromosome - assembled scaffolds

Assembly from short reads is hard

Basic Principle

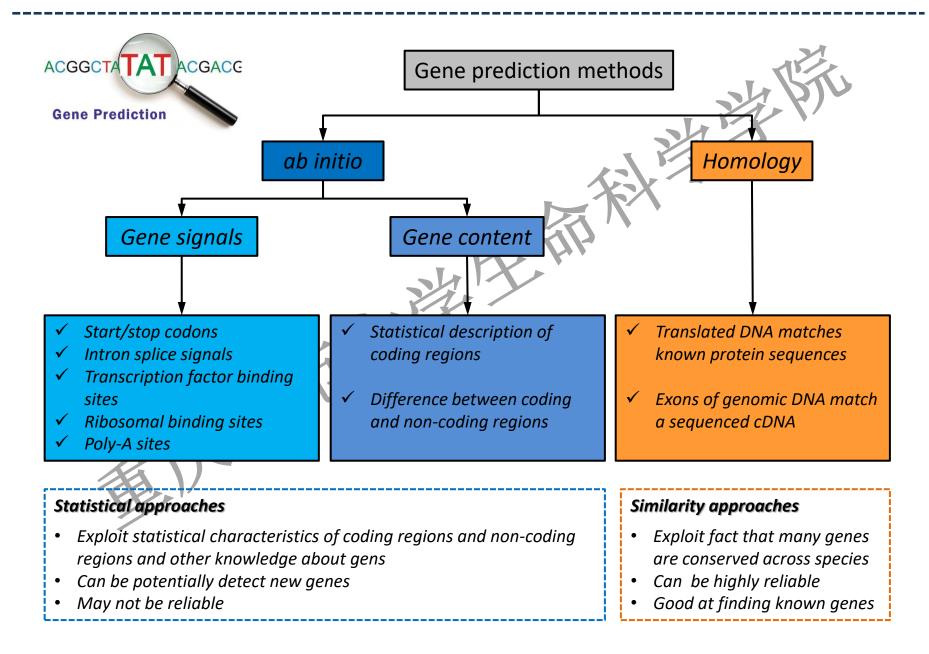




Anatomy of whole-genome assembly. Overlapping shredded bactig fragments (red lines) and internally derived reads from five different individuals (black lines) are combined to produce a contig and a consensus sequence (green line). Contigs are connected into scaffolds (red) by using mate pair information. Scaffolds are then mapped to the genome (gray line) with STS (blue star) physical map information.

(read/村 - Contig/镇 - Scaffold/县 - Chromosome/省 - Genome/国家)

◆ 4.3 基因识别(预测)



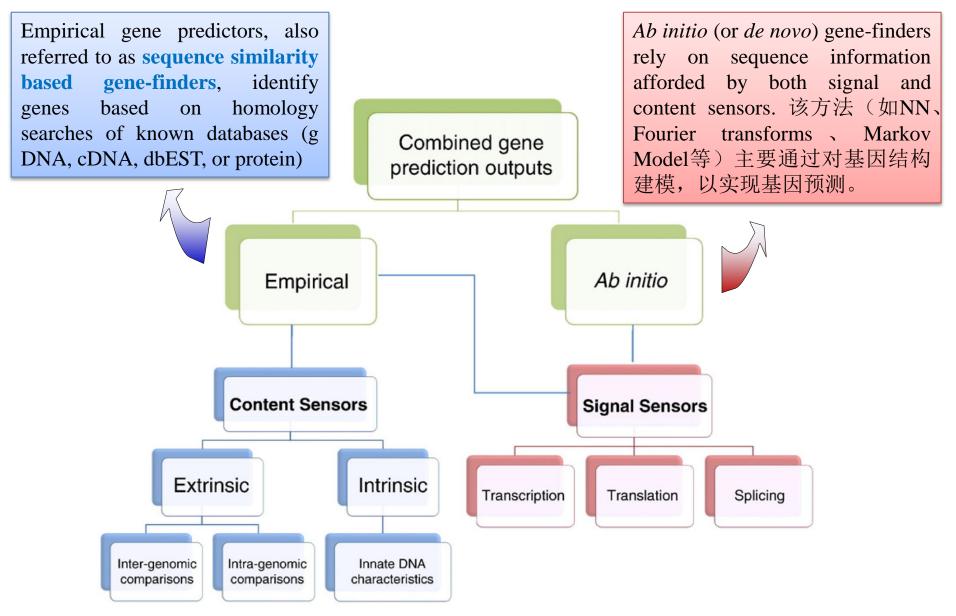
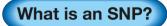
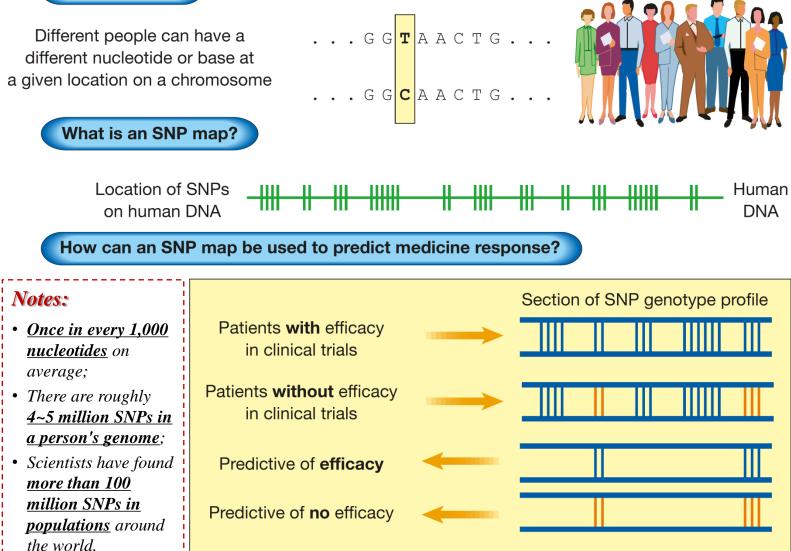


Fig. Schematic overview of eukaryote gene prediction methods and the underling sensors routinely used to locate genes in genomic sequences

Ref: Sleator R D. *Gene*, 2010, 461(1-2): 1-4.

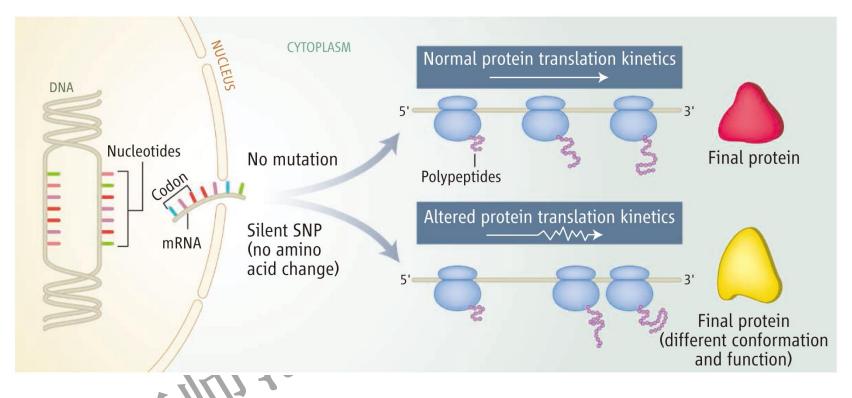
◆ 4.4 基因多交性分析(如SNP)





Roses A D. Pharmacogenetics and the practice of medicine[J]. Nature, 2000, 405(6788): 857-865.

> SNP and Protein



Translation kinetics and protein folding. Unaffected translation kinetics results in a correctly folded protein. Abnormal translation kinetics, caused by the ribosome moving faster or slower through certain mRNA regions, can produce a different final protein conformation. Abnormal kinetics may arise from a silent single nucleotide polymorphism (SNP) in a gene that creates a codon synonymous to the wild-type codon. However, this synonymous codon substitution may lead to different kinetics of mRNA (protein) translation, thus yielding a protein with a different final structure and function.

◆ 4.5 细胞内RNA表达分析

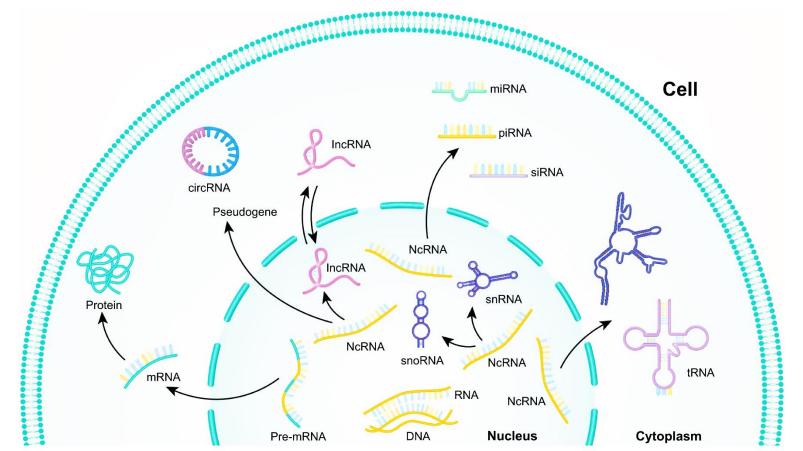


Fig. 1 The classification of coding and noncoding RNA. Eukaryotic mRNA molecules are usually composed of small segments of the original gene and are generated by a process of cleavage and rejoining from an original precursor RNA (pre-mRNA) molecule, which is an exact copy of the gene. Noncoding RNA (ncRNA) mainly include long non-coding RNA (lncRNA), microRNA (miRNA), pseudogene, circular RNA (circRNA), small interfering RNA (siRNA), piwi-interacting RNA (piRNA), small nucleolar RNA (snoRNA), small nuclear RNA (snoRNA), ribosomal RNA (rRNA) and transfer RNA (tRNA)

Xu G, Xu W Y, Xiao Y, et al. The emerging roles of non-coding competing endogenous RNA in hepatocellular carcinoma[J]. Cancer Cell International, 2020, 20(1): 1-21.

The World of RNAs **Coding RNAs** Non-coding **RNA** mRNAs Structural non-**Regulatory non**coding RNAs coding RNA tRNAs rRNAs Small non-coding Medium non-coding Long non-coding (House-keeping RNAs) **RNA RNA RNA** ncRNA: non-coding RNA, **miRNA** Structure **rRNA**: ribosomal RNA. **Biogenesis** Action **SnoRNA PiRNA** tRNA: transfer RNA, tiRNA SiRNA miRNA: micro RNA. **SnRNA** CrasiRNA piRNA: piwi RNA, CisRNA Intronic RNA **ScRNA TelsRNA** siRNA: small interfering RNA, Linear CeRNA **Enhancer RNA** PROMPTs crasiRNA: Centromere repeat associated IncRNAs TransRNA **Promoter RNA** small interacting RNA, telsRNAs: telomere-specific small RNA, eRNAs: enhancer-derived RNAs, Antisense RNA snoRNA: small nucleolar RNA, NATs: natural antisense transcript, Sense RNA Circular tiRNA: transcription initiation RNA, **TUCRNAs:** transcribed ultraconserved **Intergenic RNA** IncRNAs snRNA: small nuclear RNA, RNAS, **Bidirectional RNA** scRNA: small cytoplasmic/ conditional cis-lncRNA: cis-acting long non coding lincRNA RNA, RNA, eRNA **Prompts:** promoter-upstream transcripts, trans-lncRNA: trans-acting long non TUCRNA *lincRNAs:* Long intergenic noncoding coding RNA, NAT ceRNA: competing endogenous RNA. RNAs,

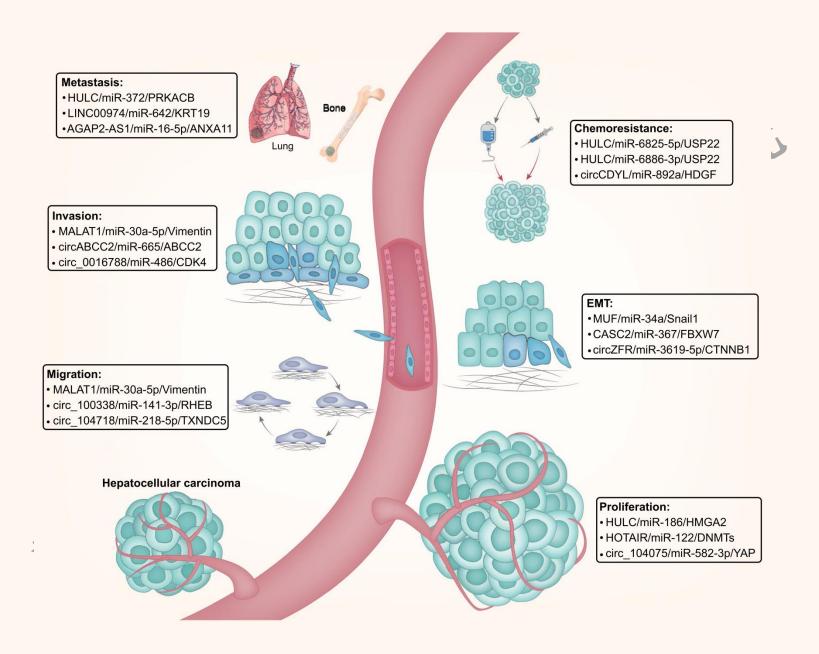
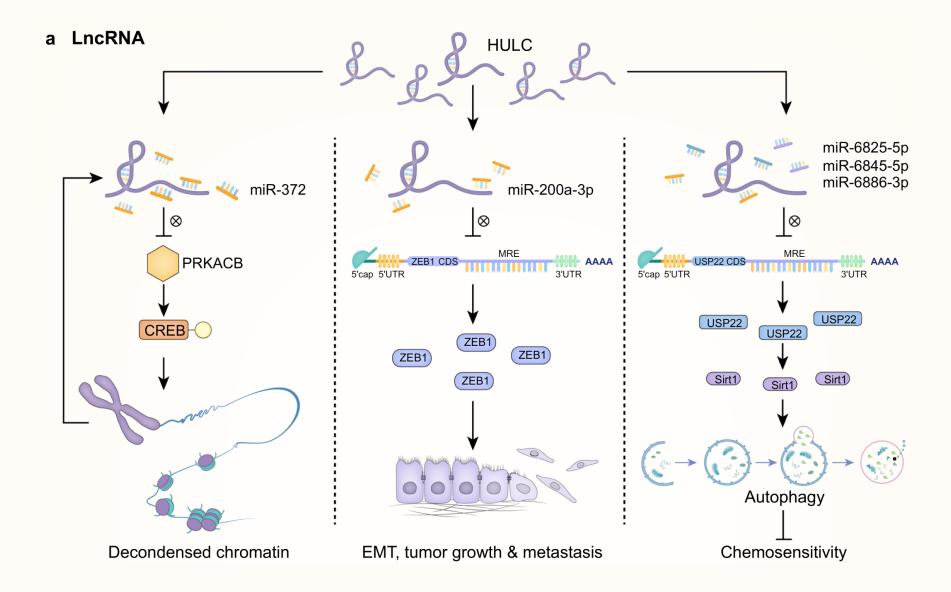
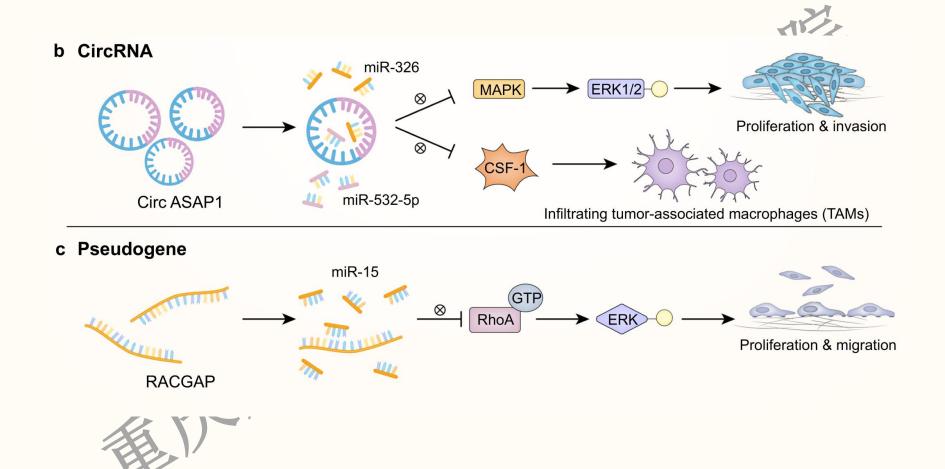


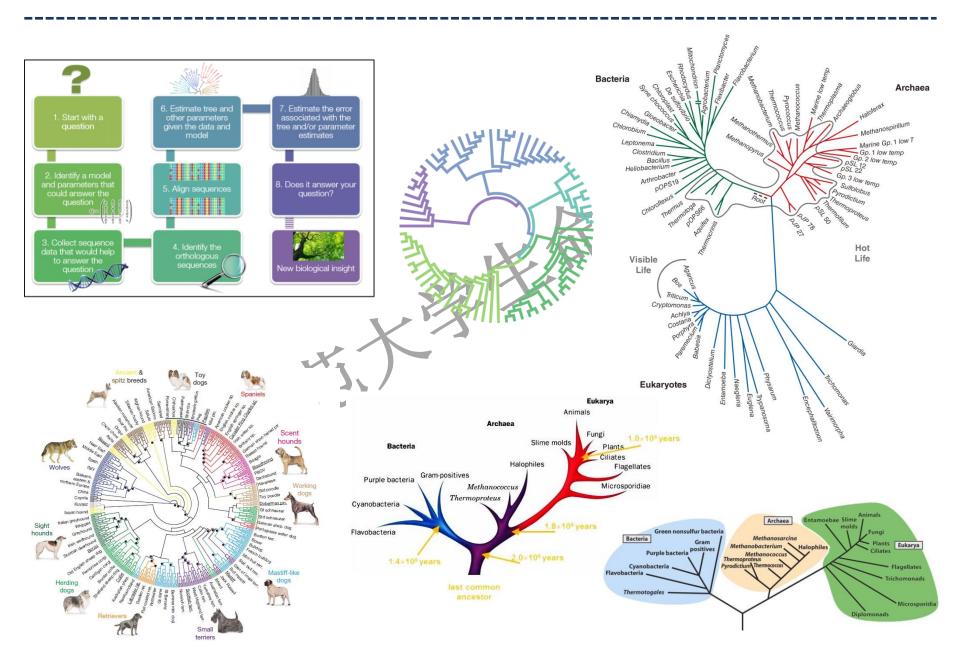
Fig. Summary of ncRNAs act as ceRNAs mediated function in hepatocellular carcinoma progression.





Xu G, Xu W Y, Xiao Y, et al. The emerging roles of non-coding competing endogenous RNA in hepatocellular carcinoma[J]. Cancer Cell International, 2020, 20(1): 1-21.

◆ 4.6分子进化分析



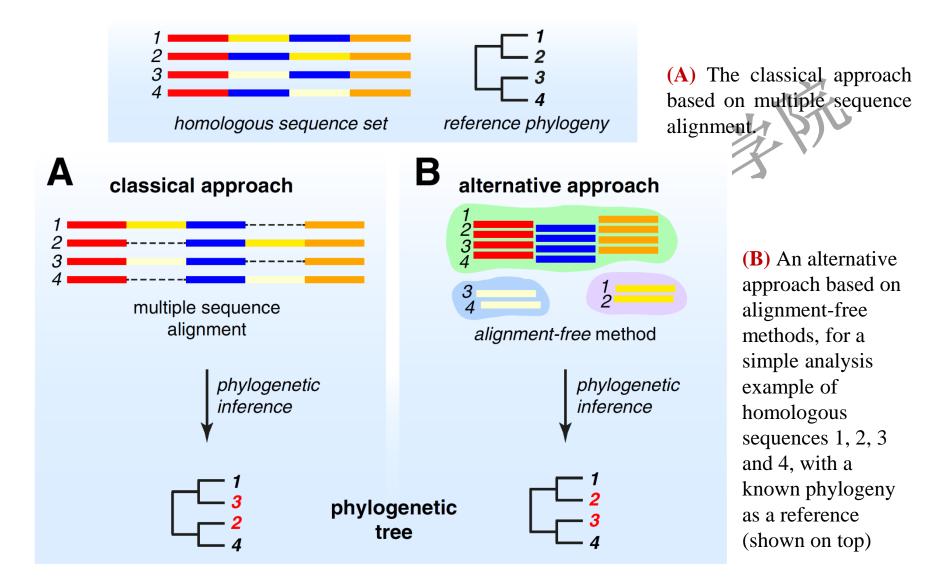
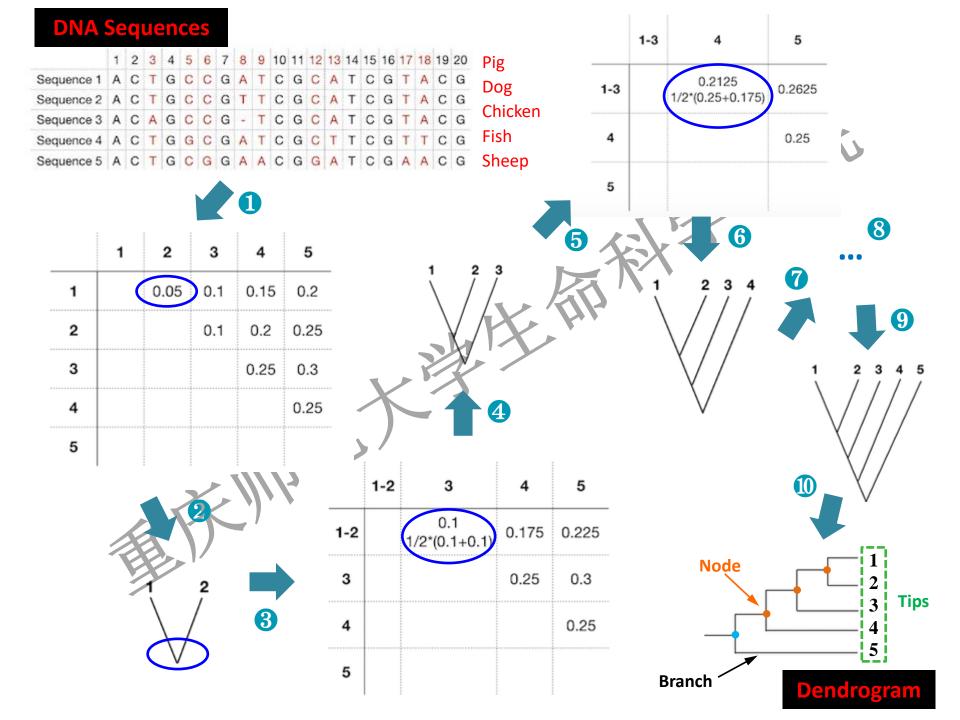
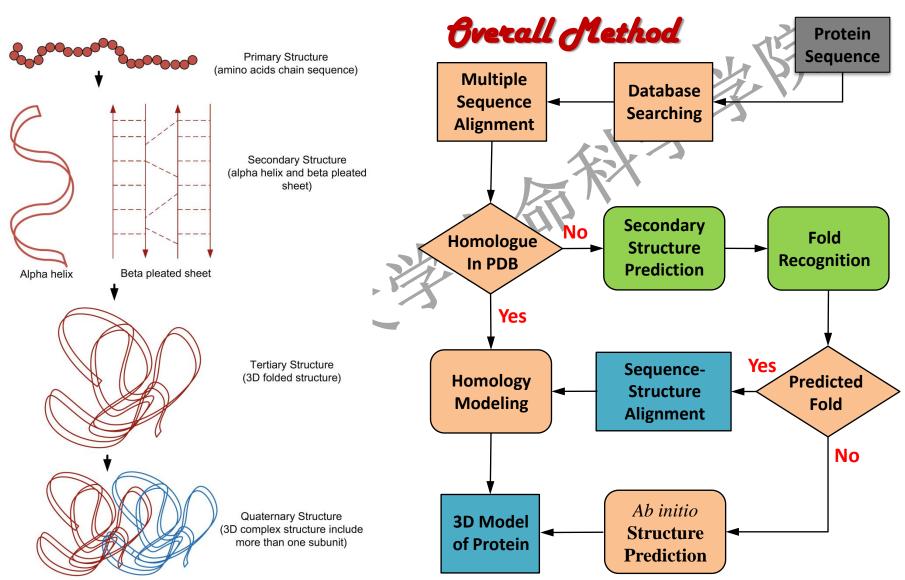


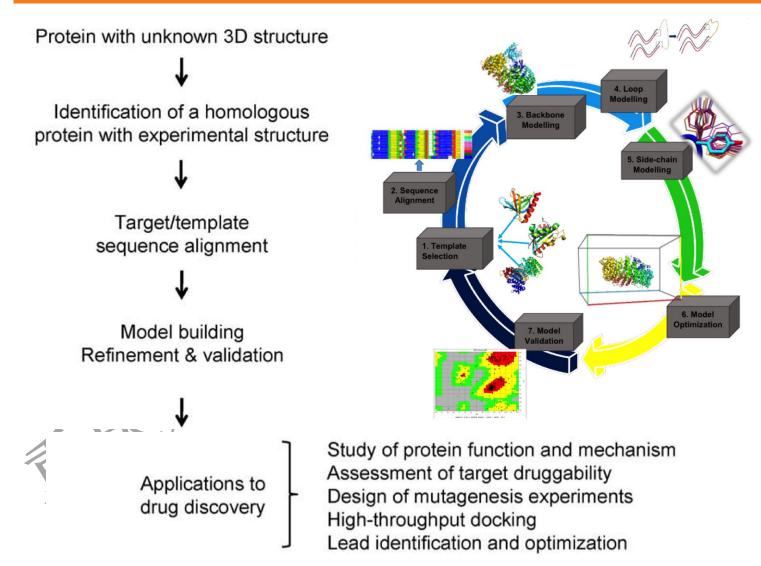
Fig. Simplified workflow of phylogenomic approaches, A and B.



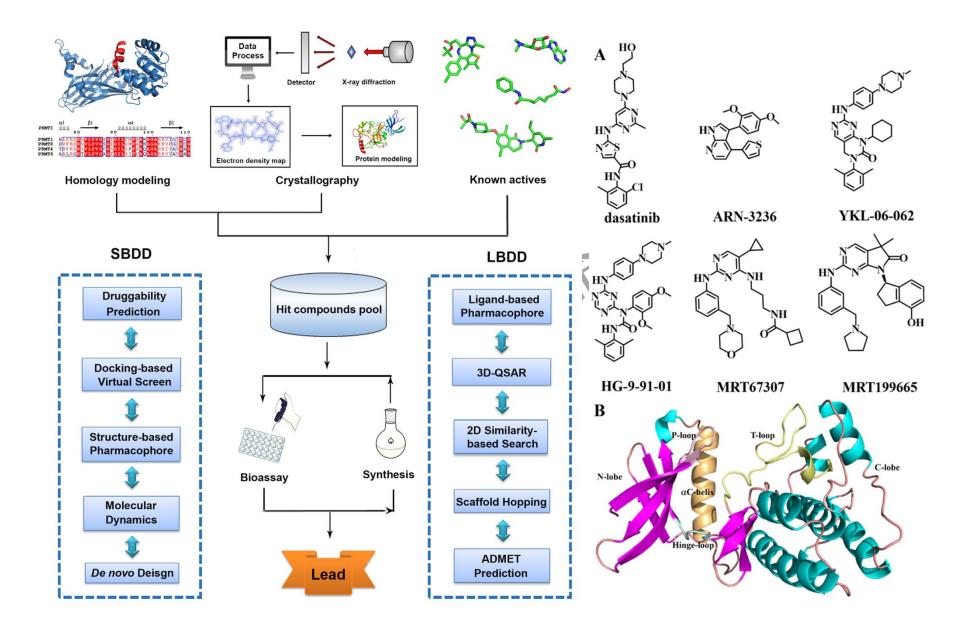




Homology modeling in drug discovery: current trends and applications

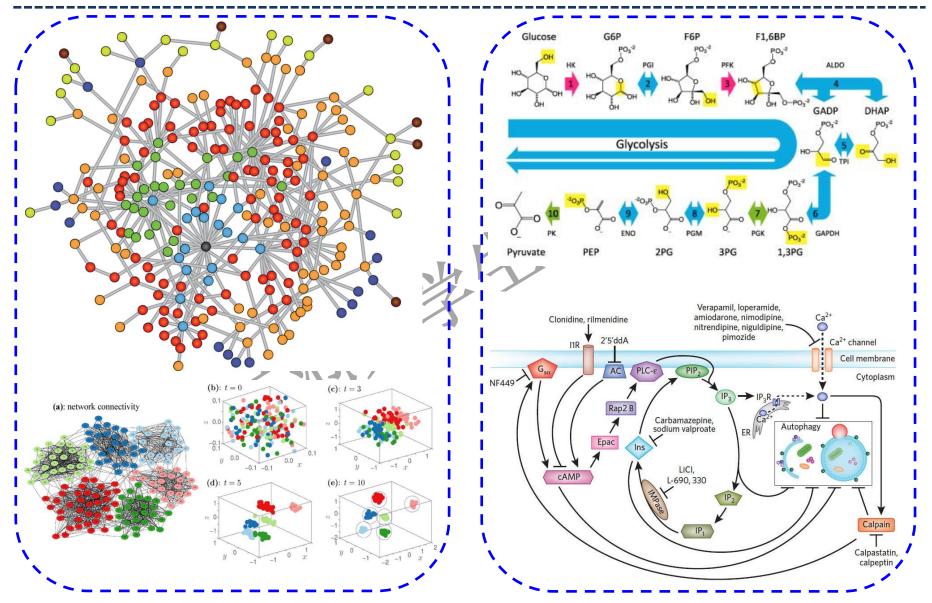


Refs: Munsamy G, et al. *Letters in Drug Design & Discovery*, 2017, 14(9): 1099-1111. Cavasotto C N, *et al*. Drug discovery today, 2009, 14(13-14): 676-683.



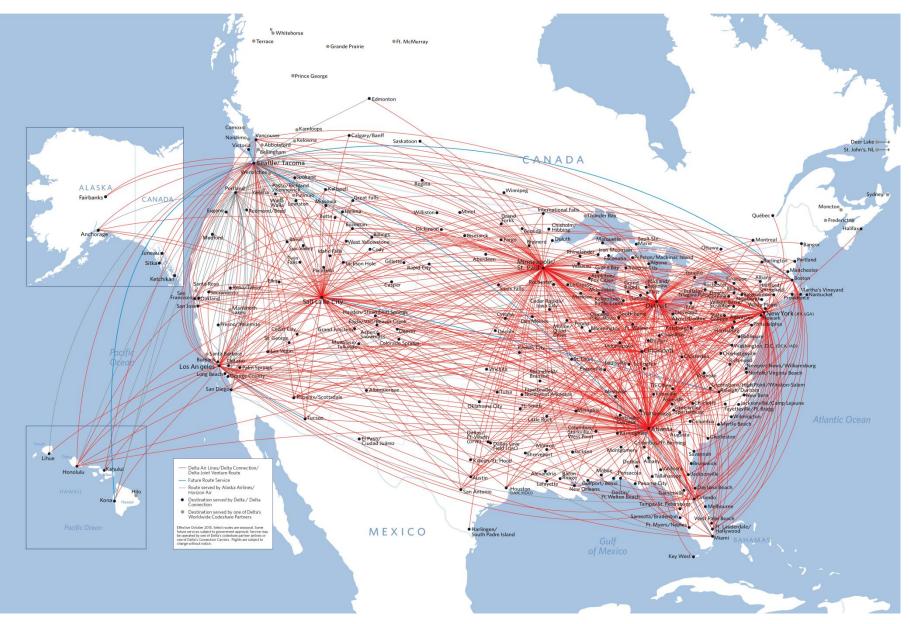
Traditional workflow of structure-based drug design (SBDD) and ligand-based drug design (LBDD).

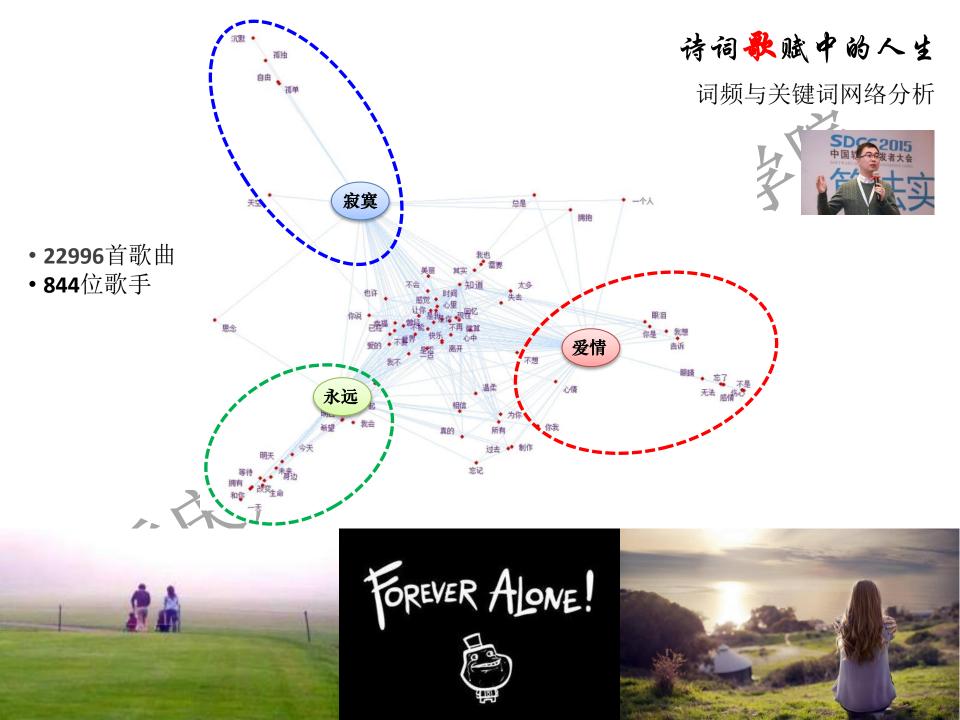
◆ 4.8 分子间相互作用

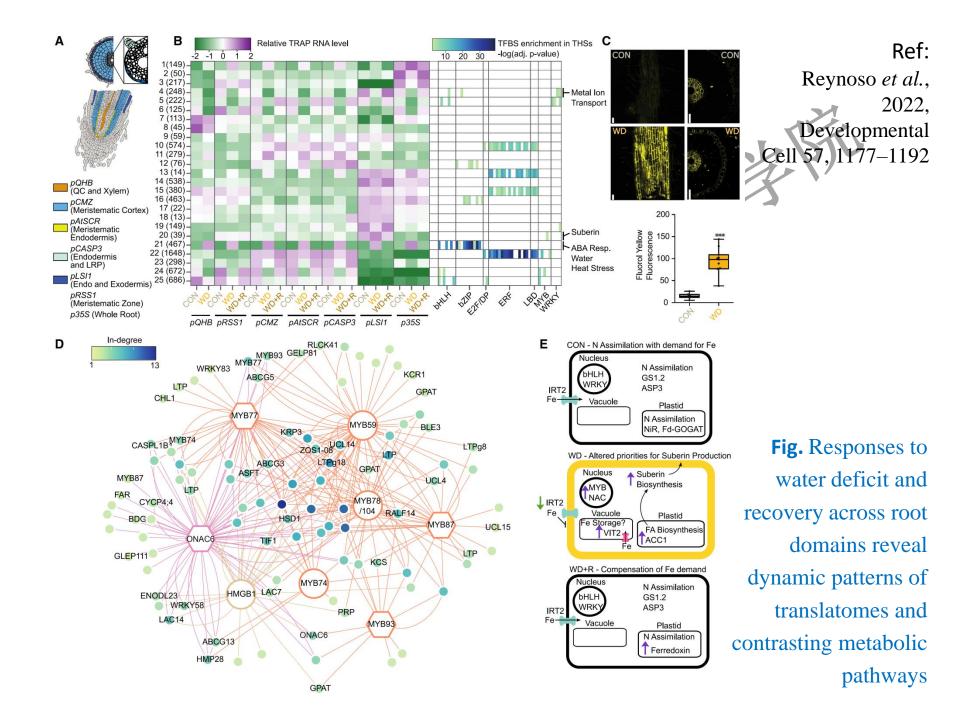


网络分析与通路分析

★美国及其周边重要航空枢纽



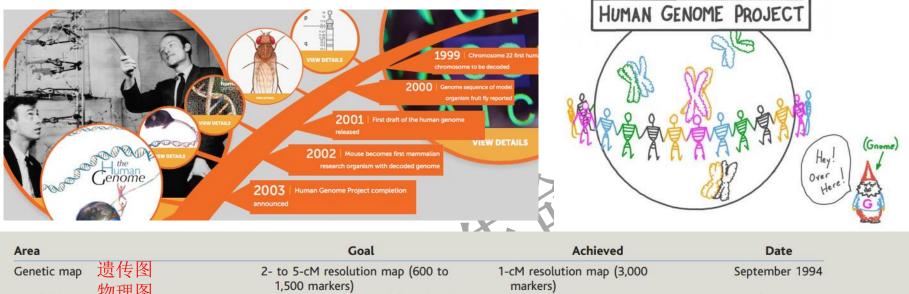








◆ 5.1人类基因组计划



| Physical map | 物理图 |
|--------------|-----|
| DNA sequence | 序列图 |

Capacity and cost of finished sequence

Human sequence variation Gene identification 基因图 Model organisms 2- to 5-cM resolution map (600 to 1,500 markers)
30,000 sequence-tagged sites (STSs)
95% of gene-containing part of human sequence finished to 99.99% accuracy
Sequence 500 Mb/year at <\$0.25 per finished base
100,000 mapped human SNPs
Full-length human cDNAs
Complete sequences of *E. coli*, *S. cerevisiae*, *C. elegans*,

D. melanogaster

Functional analysis

Develop genomic-scale technologies

The human genome project was initiated in **1990** in order to sequence the whole genetic content of the human genome and other species to know genes and their functions.

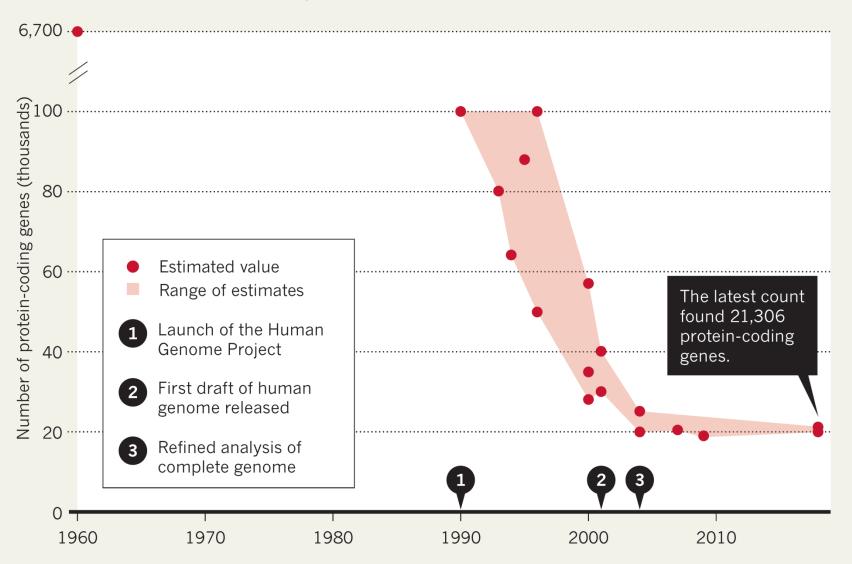
| Achieved | Date | |
|--|------------------|-------|
| 1-cM resolution map (3,000 markers) | September 1 | 994 |
| 52,000 STSs | October 199 | 8 |
| >98% of gene-containing part of human sequence finished to 99.99% accuracy | April 2003 | |
| Sequence >1,400 Mb/year at <\$0.09 per finished base | November 2 | 002 |
| 3.7 million mapped human SNPs | February 20 | 03 |
| 15,000 full-length human cDNAs | March 2003 | |
| Finished sequences of E. coli, S. cerevisiae, C. elegans, D. | April 2003 | |
| melanogaster, plus whole-genome | | TP53 |
| drafts of several others, including | | TNF |
| C. briggsae, D. pseudoobscura, | | |
| mouse, and rat | 2000 C 10 C 10 C | EGFR |
| High-throughput oligonucleotide synthesis | 1994 | IL6 |
| DNA microarrays | 1996 | VEGFA |
| Normalized and subtracted cDNA libraries | 1996 | APOE |
| Eukaryotic, whole-genome knockouts (yeast) | 1999 | TGFB1 |
| Scale-up of two-hybrid mapping | 2002 | MTHFR |

Top 8 genes

GENE TALLY

对人类基因组的认识越来越清晰

Scientists still don't agree on how many protein-coding genes the human genome holds, but the range of their estimates has narrowed in recent years.



Ref: Willyard C. New human gene tally reignites debate[J]. *Nature*, 2018, 558(7710): 354-356.



杨焕明院士(前排右二)



Ref: X. L. Wang, et al. Protein & Cell, 2018, 9(4): 317-321

RECOLLECTION

The international Human Genome Project (HGP) and China's contribution

Xiaoling Wang, Zhi Xia, Chao Chen, Huanming Yang[⊠]

BGI-China, Shenzhen 518083, China ⊠ Correspondence: yanghuanming@genomics.cn (H. Yang) 截止目前,人类基因组 被明确的基因数目如下:

| Locus Group | Locus Type | Count |
|---------------------------------|----------------------------|-------|
| protein-coding | gene with protein product | 19193 |
| gene | | |
| (19193) 19206 † | | |
| non-coding | RNA, Y | 4 |
| RNA | RNA, cluster | 119 |
| (8581) 8906 ↑ | RNA, long non-coding | 5243 |
| | RNA, micro | 1912 |
| | RNA, misc | 30 |
| | RNA, ribosomal | 60 |
| | RNA, small nuclear | 50 |
| | RNA, small nucleolar | 568 |
| | RNA, transfer | 591 |
| | RNA, vault | 4 |
| Pseudogene | T cell receptor pseudogene | 36 |
| (13908) 14008 † | immunoglobulin pseudogene | 203 |
| | pseudogene | 13669 |
| Other | T cell receptor gene | 201 |
| (1035) 1004 † | complex locus constituent | 29 |
| | endogenous retrovirus | 109 |
| | fragile site | 116 |
| | immunoglobulin gene | 229 |
| | protocadherin | 39 |
| | readthrough | 138 |
| | region | 38 |
| | unknown | 128 |
| | virus integration site | 8 |

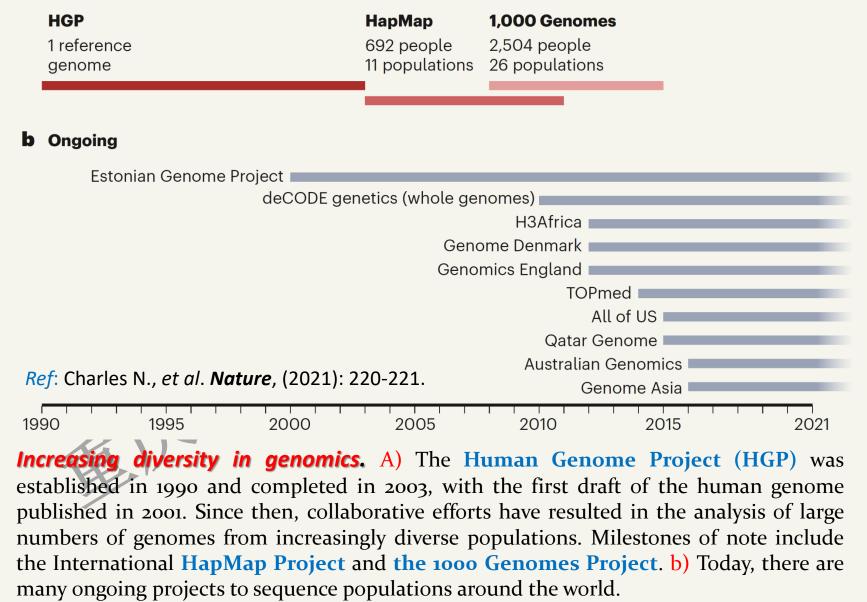
Total Approved (42717) 43124 †

Last update: 17/08/21 Newly update: 21/08/22

https://www.genenames.org/download/statistics-and-files/

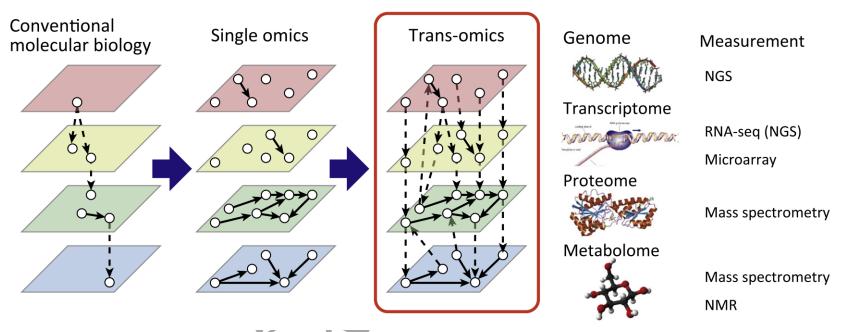
多样化的人类基因组

a Milestones

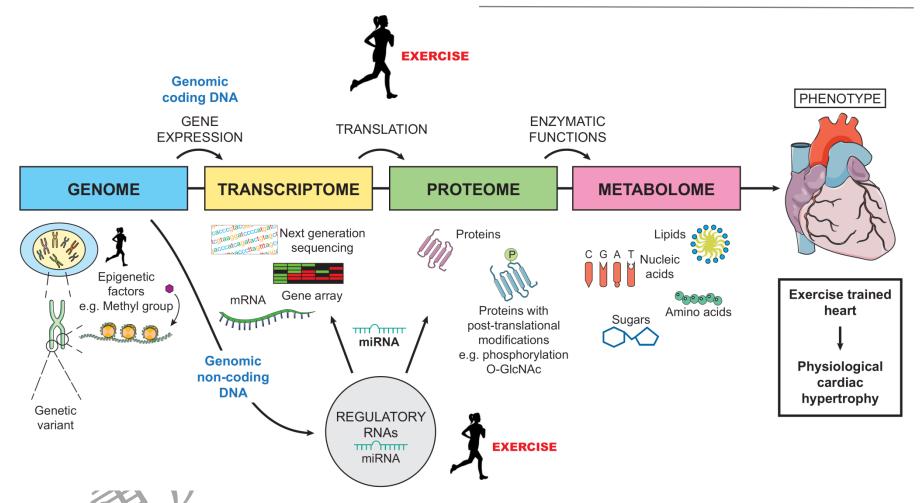


◆ 5.2 生物组学 (Omics)

Ref: Yugi K, *et al*. <u>Trends in biotechnology</u>, 2016, 34(4): 276-290.



- ✓ What can happen? Human genome contains roughly 3 billion nucleotides and just under 20,000 protein-coding genes an estimated 1% of the genome's total length.
- ✓ What appears to be happening? Approximately 360,000 mRNA molecules are present in a single mammalian cell, made up of about 12,000 (14,000 for human) different transcripts with a typical length of around 2 kb. Some mRNAs comprise 3% of the mRNA pool whereas others account for less than 0.1%. These rare or low-abundance mRNAs may have a copy number of only 5~15 molecules per cell.
- ✓ What makes it happen? Human body contains 80,000~400,000 proteins in proteome, while A typical cell holds 42 million protein molecules, scientists reveal.
- ✓ What actually happens? HMDB collects detailed information ~3100 metabolites found in human urine along with 4651 metabolites found in human serum.



The use of multi-omics platforms to identify novel mechanisms and uncover exercise signatures. Integrating data from multi-omics systems to understand <u>genetic variants and</u> <u>epigenetic marks</u>, <u>gene expression and miRNAs</u>, <u>proteins</u>, and <u>metabolites</u> during exercise to define molecular pathways of exercise.

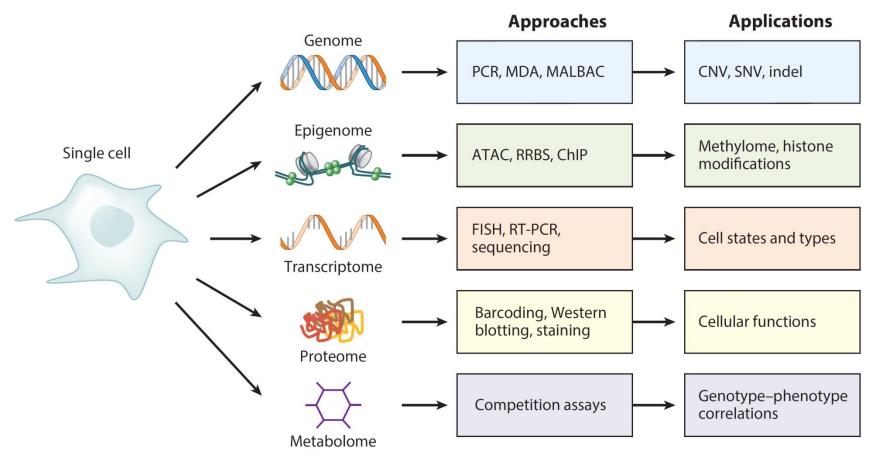
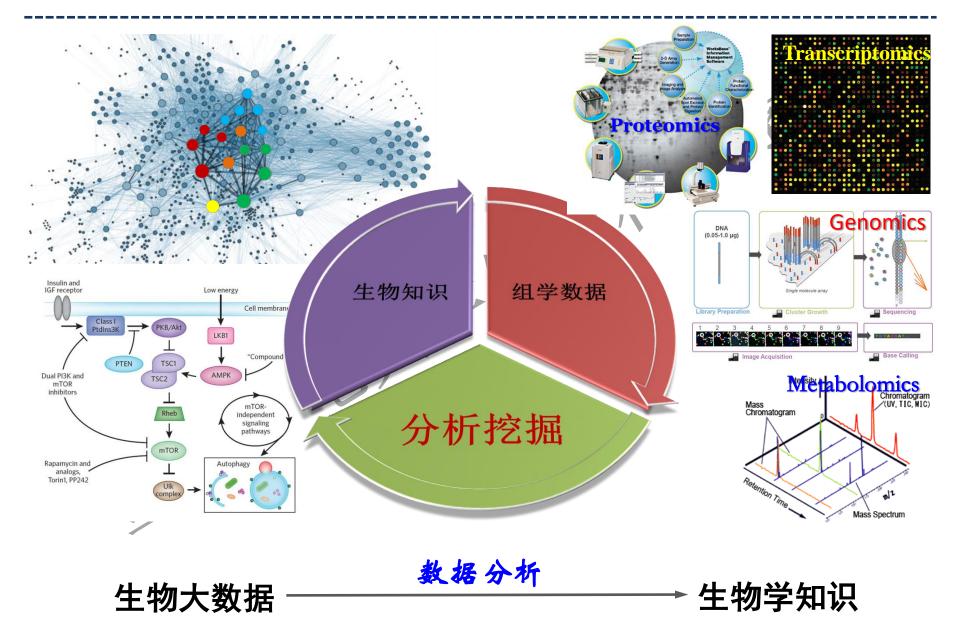


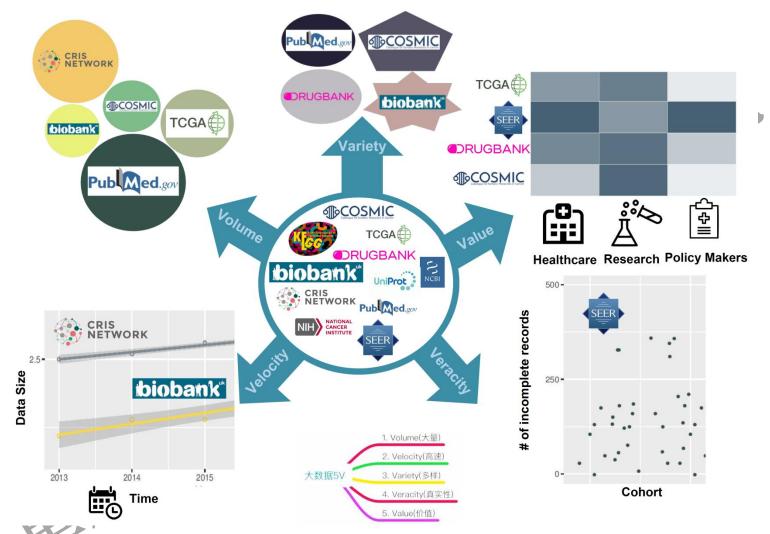
Figure 1

Ref: Deng Y, et al. Annual review of biomedical engineering, 2019, 21: 365-393.

Overview of approaches and applications in single-cell omics measurement. Abbreviations: ATAC, assay for transposase-accessible chromatin; ChIP, chromatin immunoprecipitation; CNV, copy number variation; FISH, fluorescence in situ hybridization; indel, insertion/deletion; MALBAC, multiple annealing and loop-based amplification cycling; MDA, multiple displacement amplification; PCR, polymerase chain reaction; RRBS, reduced-representation bisulfite sequencing; RT-PCR, reverse transcription polymerase chain reaction; seq, sequencing; SNV, single-nucleotide variant.

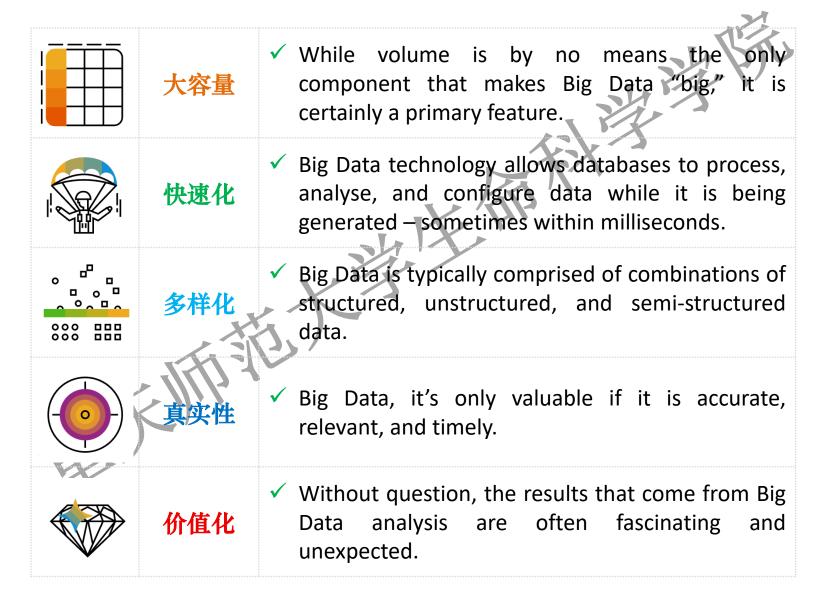
◆ 5.3 生物大数据与知识发现

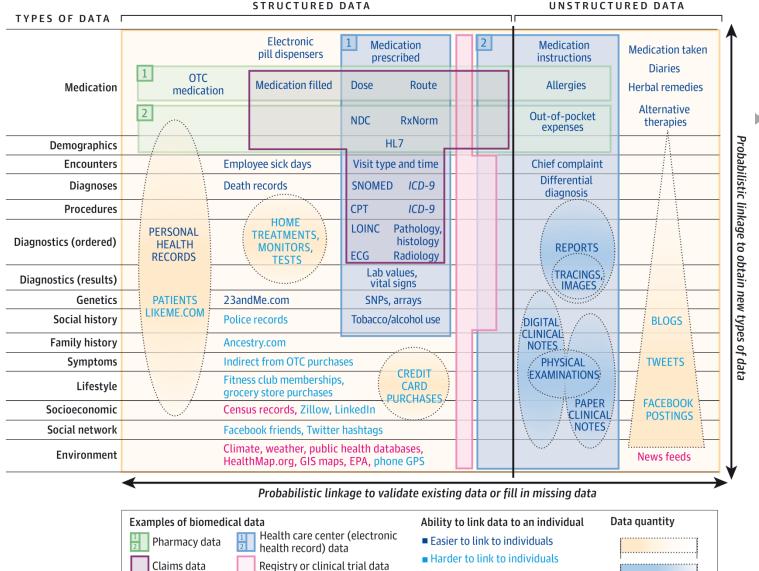




Big Biomedical Data. The **5Vs model** is utilized to characterize the very nature of big biomedical data. As observed, the dominant big data dimensions, i.e., volume, velocity, variety, veracity, and value, are present in existing biomedical datasets.

□ 大数据的3V~8V





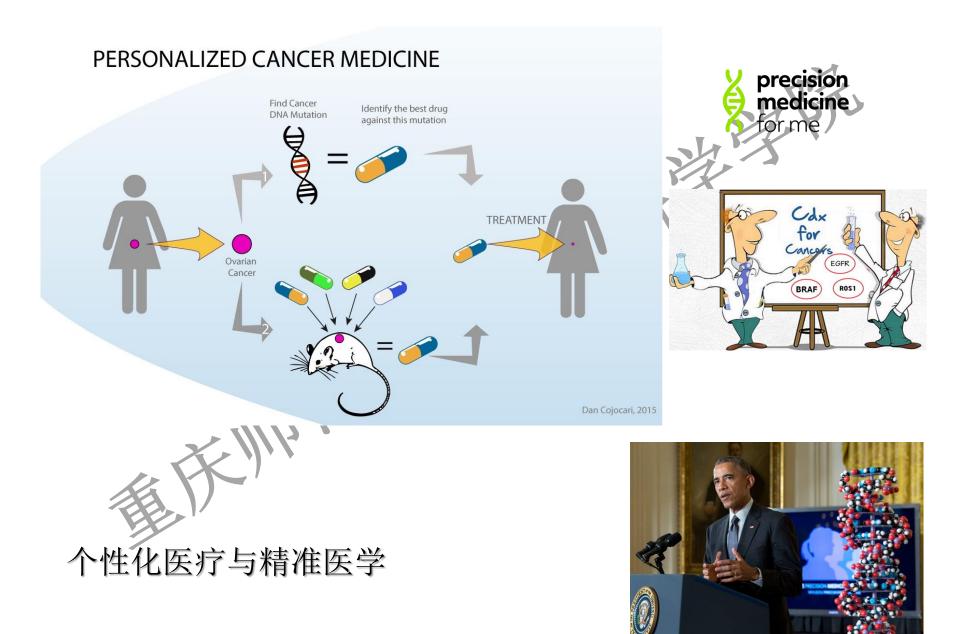
Only aggregate data exists

Less

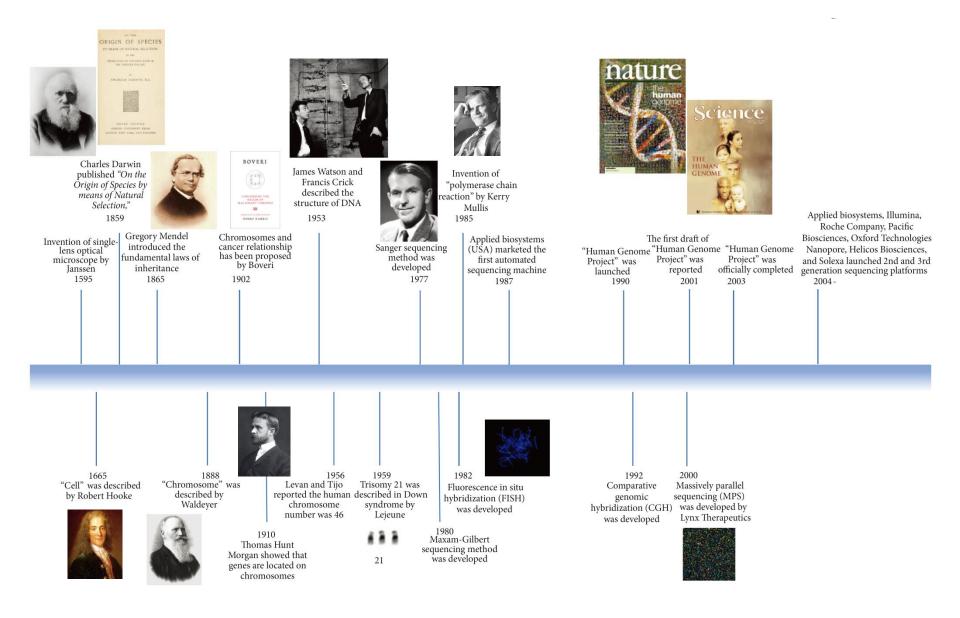
More

Fig: 潜在的高价值信息来源的图谱,可能与个人联系在一起用于医疗保健

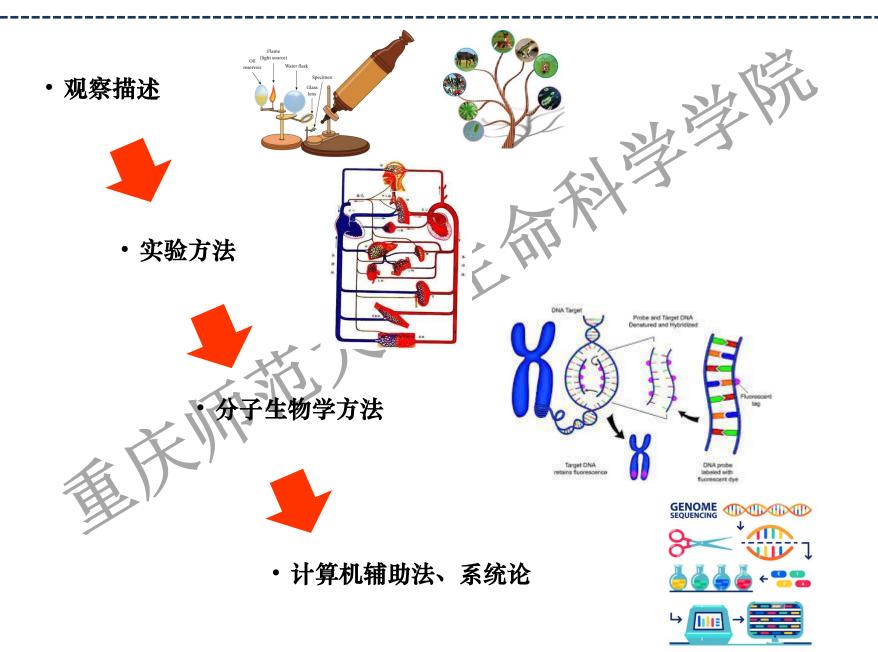
Data outside of health care system

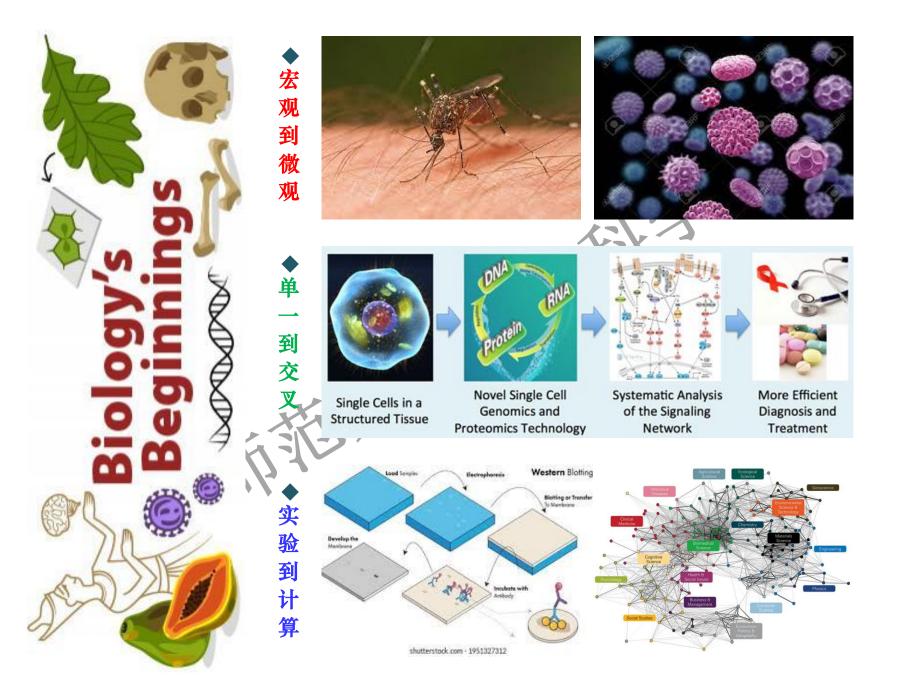


★* 第6节:总结与展望

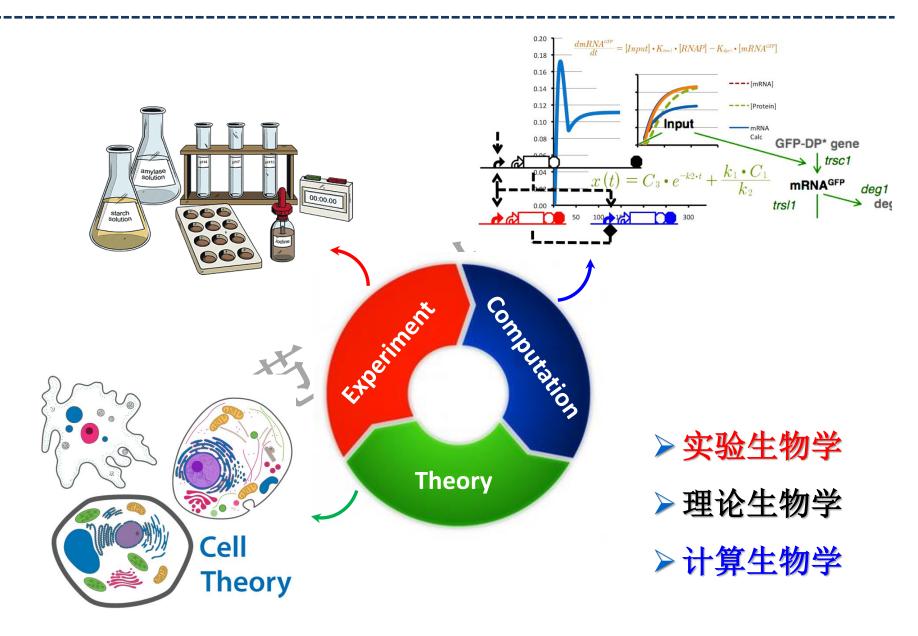


◆ 6.1 生物学发展简史及特点





◆ 6.2 生物学知识的来源各异





运用计算机科学的基本理念,进行问题求解、系统设计及理解人类 行为。即一种运用计算机科学的基本理念来解决问题的思考方式。

The Computational Thinkers

concepts



Logic Predicting & analysing

Evaluation Making judgements

0000

Algorithms Making steps & rules



Patterns Spotting & using similarities



Decomposition Breaking down into parts



Abstraction Removing unnecessary detail



We are all computational thinkers here!





Chanks for your attention!

Acknowledgement

College of Life Sciences, Chongqing Normal University 2022, Chongqing of P. R. C